

Lecture 8

Plan:

- Faces of Polyhedra
- State tons of facts
- Prove them

Faces of Polyhedra

Def: $a^{(1)}, \dots, a^{(k)} \in \mathbb{R}^n$ are

affinely independent if

$$\sum_{i=1}^k \lambda_i a^{(i)} = 0$$

and $\sum \lambda_i = 0$ imply $\lambda_1 = \dots = \lambda_k = 0$.

(w/out $\sum \lambda_i = 0$, is just linear indp.)

linear independent \Rightarrow affine independent.

Note:

$\{c(i)\}$ n.o. ... dependent iff

$\left\{ \alpha \rightarrow \text{affinely independent} \right.$

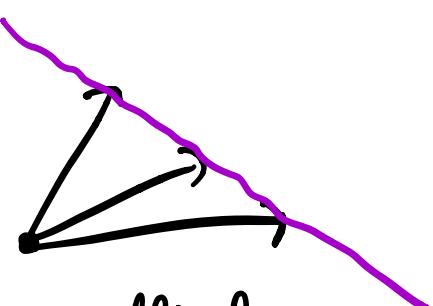
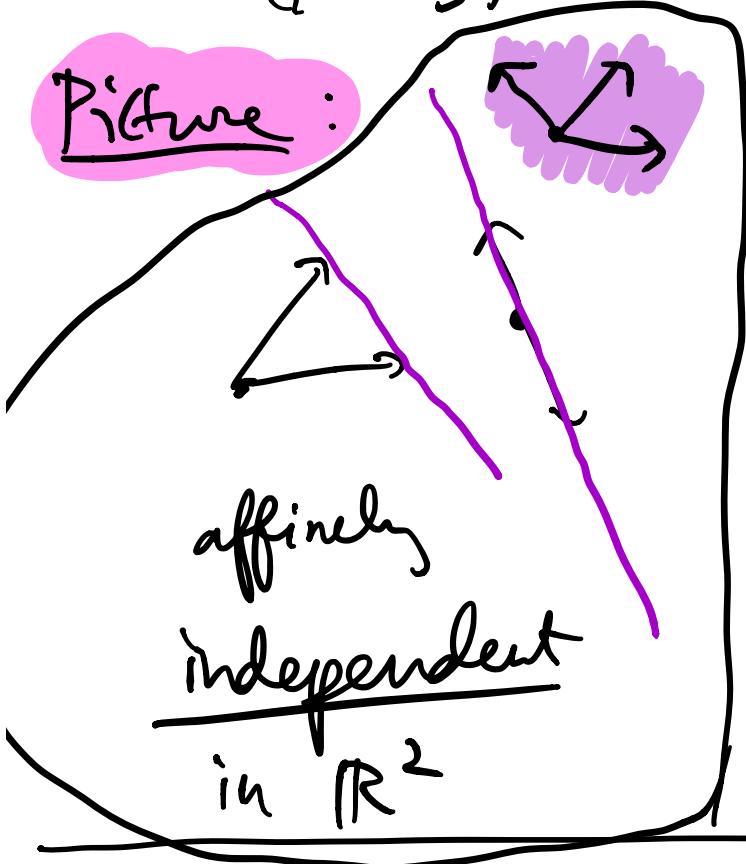
$$\left\{ \begin{bmatrix} \alpha^{(i)} \\ 1 \end{bmatrix} \right\}$$

linearly independent.

$\Leftrightarrow \text{aff}\left(\{\alpha^{(i)}\}\right)$ has dimension $K-1$

vectors.

Picture:



affinely
dependent
in \mathbb{R}^2 .

Def Dimension $\dim(P)$ of

polyhedron P :

$-1 + \max \# \text{affinely}$
 $\text{independent points in } P.$

Equivalently, dimension of
affine hull $\text{aff}(P)$.

Example: $P = \emptyset, \dim(P) = -1$

$P = \text{singleton} \quad \cdot \quad \dim(P) = 0$

$P = \text{line segment} \quad \nearrow \quad \dim(P) = 1$

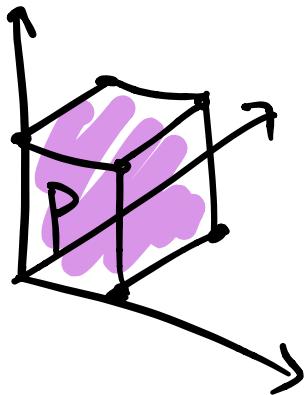
;

$$\text{aff}(P) = \mathbb{R}^n$$

$$\dim(P) = n;$$

P "full dimensional"

e.g. cube in \mathbb{R}^3 : $\{x : 0 \leq x_i \leq 1\}$



$$\dim P = 3$$

$$\dim \mathbb{R}^3 = 3$$

(as polyhedron).

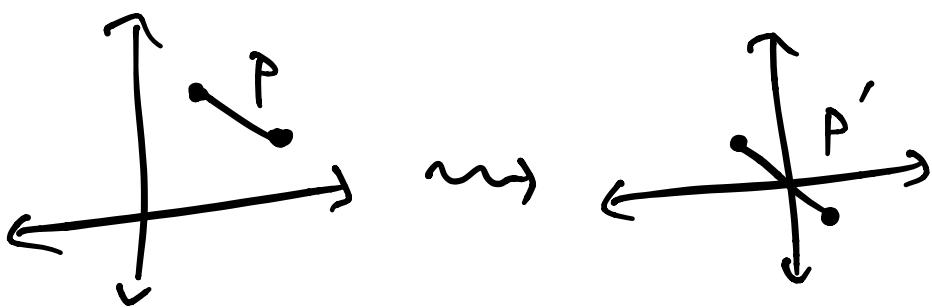
Why affine, not linear? affine

independence is translation
invariant:

if I used max # lin indep points - 1

$$\dim(P) = 1$$

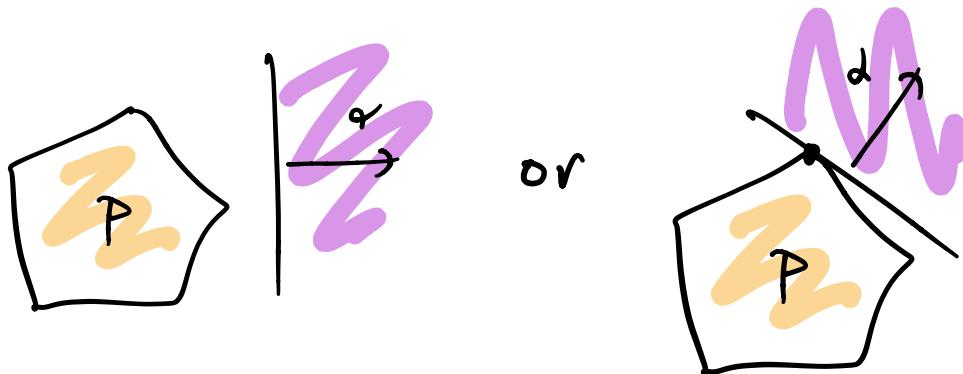
$$\dim(P') = 0$$



$$l = \dim(P) = \dim(P').$$

Def: $\alpha^T x \leq \beta$ is a valid inequality

for P if $\alpha^T x \leq \beta$ for all $x \in P$.

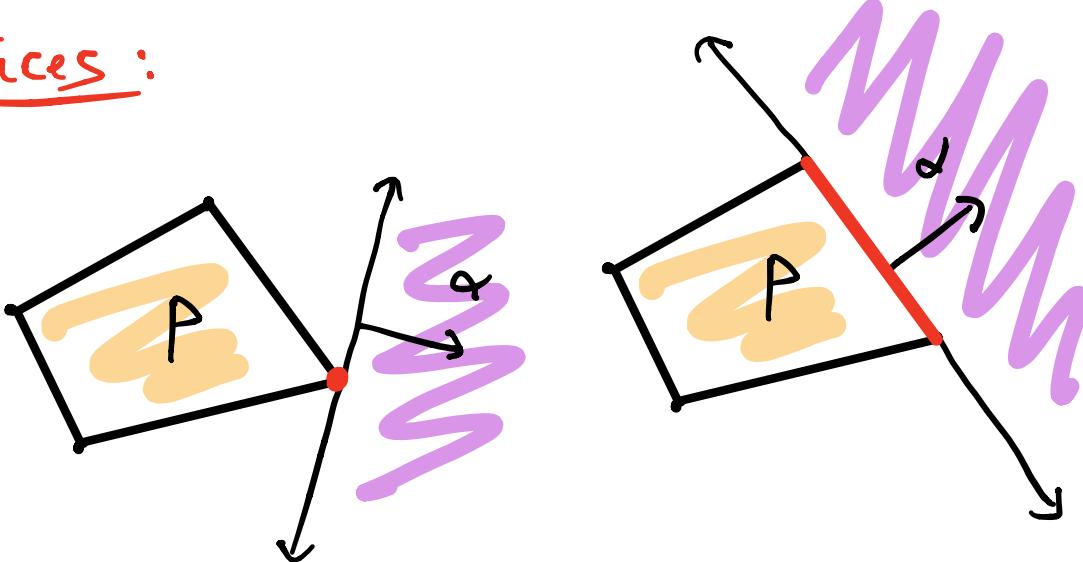


Def A face of a polyhedron

P is $\{x \in P : \}$ for

$Q^T x \leq \beta$ valid.

Faces:



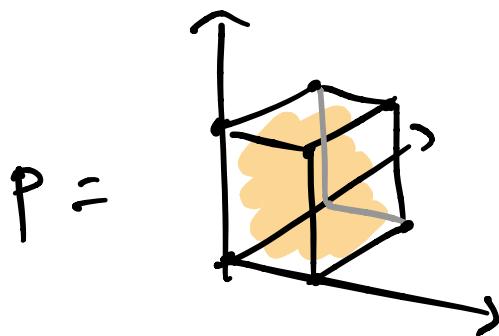
Properties:

- Faces are polyhedra
- Empty face & entire P
are called trivial faces)
- else F nontrivial
 $\leq \dim(F) \leq$
- $F : \dim(F) = \dim(P) - 1$ called facets.
in \mathbb{R}^n called vertices

~~• T - array for unitary vectors~~

Ex : list the 28 faces of the cube

$$P = \{x \in \mathbb{R}^3 : \quad \}$$



Fact : ∞ many valid ineqs,
but # faces finite!

EVERYTHING ABOUT POLYHEDRA

Let $A = \begin{bmatrix} - & a_1 & - \\ : & & \\ - & a_m & - \end{bmatrix} \in \mathbb{R}^{m \times n}$

Face Characterization!

Any nonempty face of $P = \{x : Ax \leq b\}$

is

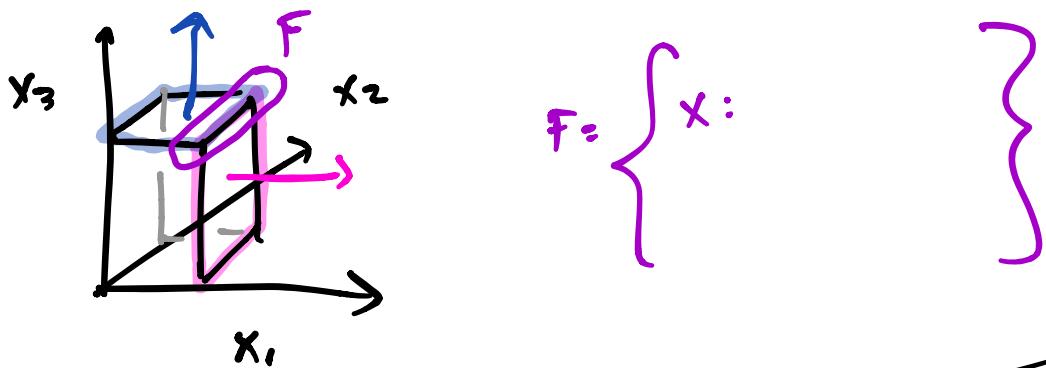
$\{x : A_I x \leq b_I\}$

$\{x : A_{I^c} x \leq b_{I^c}\}$

for some set $I \subseteq \{1, \dots, m\}$.

E.g.

cube



② Facet Maximality: The facets are the maximal nontrivial faces of a nonempty polyhedron P .

For vertices:

③ Vertex Characterization:

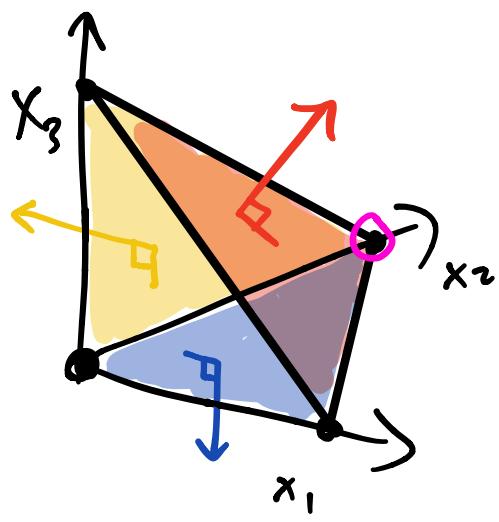
Suppose x^* extreme point of $P = \{x : Ax \leq b\}$

Then $\exists I \subseteq S$ s.t. x^* is
the unique soln to



moreover, any such unique solution $x^* \in P$
is extreme.

e.g. simplex $(0,1,0)$ is intersection of
3 constraints



- Vertex minimality: For $\text{rank}(A) = n$, minimal nontrivial faces of polyhedron

$$P = \{x \in \mathbb{R}^n : Ax \leq b\}$$

are the vertices.

- Polytopes = convex hulls

If a polyhedron P is bounded then $P = \text{conv}(\{\text{extreme points of } P\})$.
(special case of Krein-Milman theorem.)

- Facets Characterize

□ inequality $a_i^T \leq b_i$; redundant if P

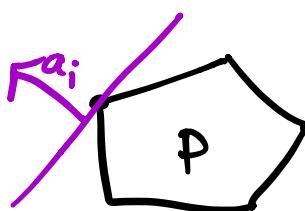
unchanged when it's removed.

□ $I_0 := \{i : a_i^T x = b_i \text{ } \forall x \in P\}$

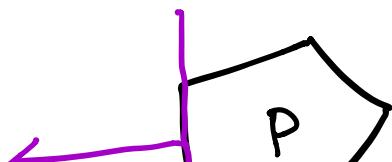
□ $I_< := \{i : \exists x \in P \text{ } a_i^T x < b_i\}$.

THEN:

(Sufficiency:) If face $a_i^T x \leq b_i$ for $i \in I_<$ is not facet, then $a_i^T x \leq b_i$ is redundant.

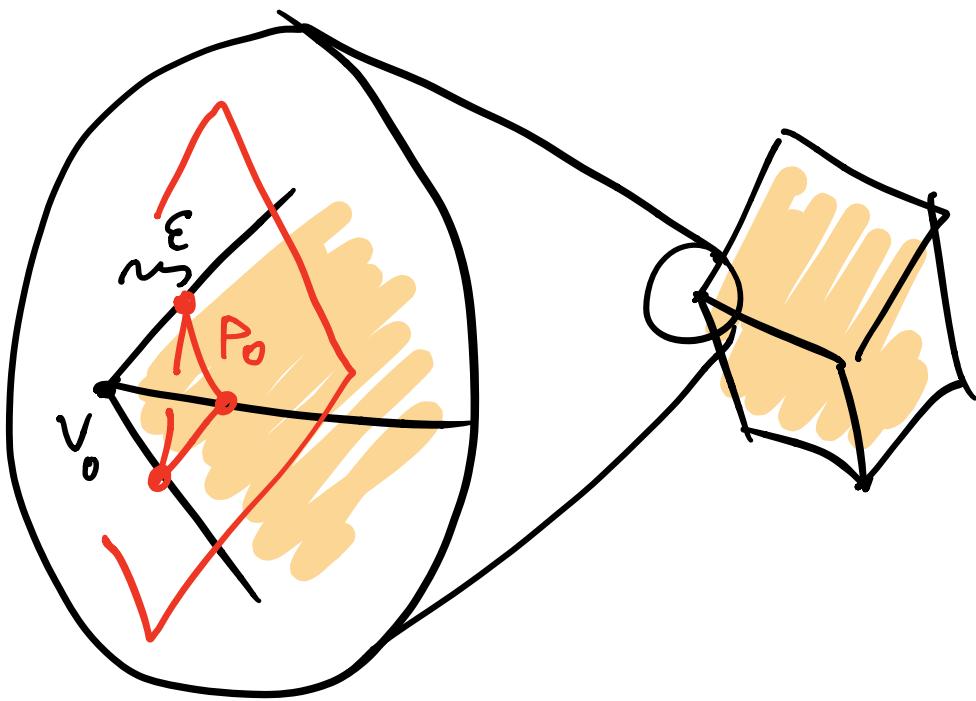


(Necessity:) If F is facet of P, $\exists i \in I_<$ such that F is induced by $a_i^T \leq b_i$.



$\sim \alpha_i$

Near vertices =
Cones over polytopes



Let v_0 vertex of P from
valid inequality $C^T x \leq m$.

Let ε be such that $c^T v' < m - \varepsilon$

for all other vertices v' .

Then

$$P_0 = \{x \in P : C^T x = m - \varepsilon\}$$

is a polytope & is bijection

$\{P_0$'s dim k faces $\}$



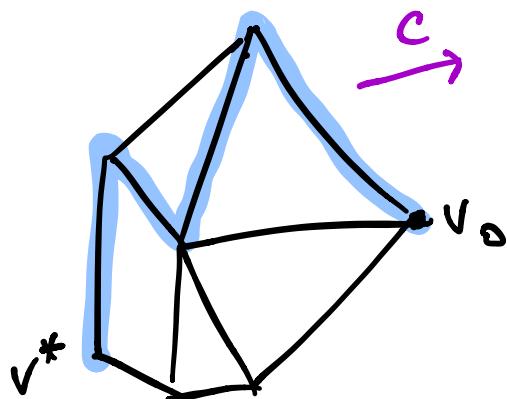
$\{P$'s dim K faces
containing $v_0\}$.

P's "graph" connected: Graph of

vertices & edges of polyhedron P

is always connected.

In particular: if v^* minimum of $c^T x$ over P ,
⇒ path from any v_0 to v with decreasing
objective along path.



PROOFS

Recall face characterization:

Let $A \in \mathbb{R}^{m \times n}$,

$$A = \begin{bmatrix} & \vdots \\ - & a_1^T & - \\ & \vdots \end{bmatrix}$$

Any nonempty face of $P = \{x : Ax \leq b\}$

is

$$\left\{ \begin{array}{l} x \\ : \end{array} \right.$$

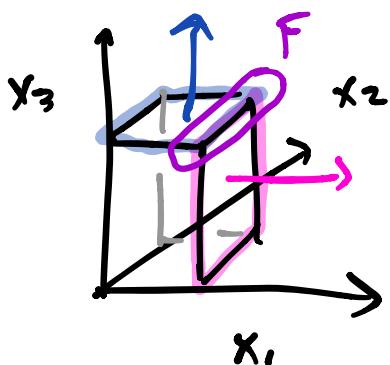
$$\left\{ \begin{array}{l} \\ \\ \\ \\ \end{array} \right.$$

for some set $I \subseteq \{1, \dots, m\}$.



E.g.

cube



$$F = \left\{ \begin{array}{l} x \\ : \end{array} \right\}$$



Proof

Consider valid inequality

$a^T x \leq b$ giving nonempty face F .

- $F = \underline{\text{optimum solutions}}$ to bounded LP

(P) \max
subject to

- Let y^* optimal solution to dual.

- Complementary Slackness:

optimal solns F are

{ $x :$

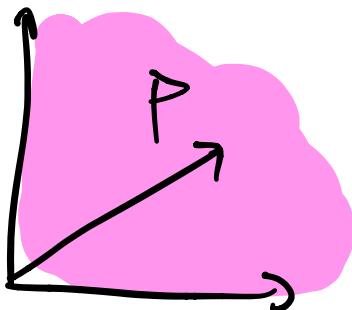
3.

Thus we can take $I = \{i : y_i^* > 0\}$. \square

Ex :

positive orthant $\{x \in \mathbb{R}^n : x_i \geq 0\} = \mathbb{P}$
has $2^n + 1$ faces

- How many of $\dim \mathbb{P}$?



For polytopes can also bound
faces in terms of # vertices.
"upper bound theorem"
Dehn-Sommerville equation

Facet minimality:

Pf : Exercise to prove from face theorem.

Recall vertex characterization:

Let x^* extreme point for

$$A = \begin{bmatrix} \vdots & \vdots \\ -a^T & - \\ \vdots & \vdots \end{bmatrix}$$

$$P = \{x : Ax \leq b\}.$$

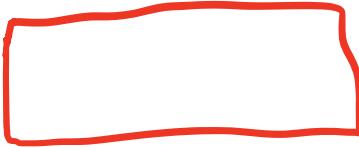
Then $\exists I$ s.t. x^* is the unique soln to



moreover, any such unique solution x^* is extreme.

Proof: Given extreme point x^* ,

- define $I = \{i : \text{ } \}$.

- Note for $i \notin I$, .

- By "faces theorem", x^* uniquely defined by

(*)

$i \in I$

(**)

$i \notin I$.

- Suppose \exists other soln. \hat{x} to (*).

- Because

 for $i \notin I$,



still satisfies $(*)$, $(**)$ for

- Contradicts F having only one point. \square .
-

Basic Feasible Solutions:

For $P = \{ \quad , \quad \}$

can describe extreme points
very explicitly.

(

).

Corollary of Vertex Thm: Extreme pts. of

$P = \{ \quad \}$ come from setting

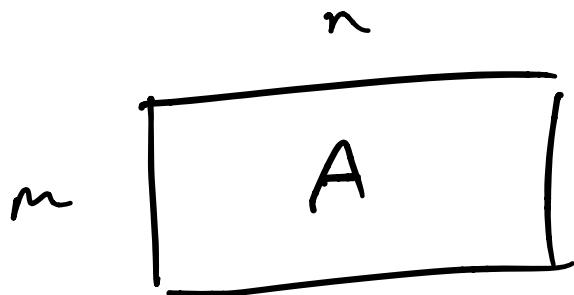


and finding unique solution to $=$
for remaining variables.

Can say more: Extreme points

of $P = \{x : Ax = b, x \geq 0\}$ are
the basic feasible solutions (BFS),
feasible solns obtained as follows:

- Remove redundant rows
from A (\rightarrow)



- Choose m columns of A , C

$$\begin{matrix} & n \\ m & \boxed{} \end{matrix} \quad \parallel \quad = \begin{bmatrix} b \end{bmatrix}$$

- Solve $A_B X_B = 0$,
set

$$x_i^* = \begin{cases} & i \in B \\ & \text{else} \end{cases}$$

$$\{ \text{bfs} \} = \{ \text{extreme pts} \}.$$

Recall vertex minimality

If $\text{rank } A = n$, vertices are minimal nontrivial facets.

Remark: Rank condition necessary; otherwise P has no vertices! (Exercise.)

Proof: Let F min'l face of P .

• Face characterization $\Rightarrow \exists I$

$$r \quad \tau \quad 1 \quad \dots \quad = \gamma$$

$$F = F_I = \left\{ x : \begin{array}{l} a_i^T x = b_i \quad \forall i \in I \\ a_j^T x \leq b_j \quad \forall j \notin I \end{array} \right\}$$

and adding any elt to I makes F_I empty.

- Consider two cases:

(a) Inequalities not needed:

$$F = \left\{ x : a_i^T x = b_i \text{ for } i \in \bar{I} \right\}.$$

* Claim: $\forall j \in I$,
 $a_j \in \text{lin}(a_i : i \in \bar{I})$.

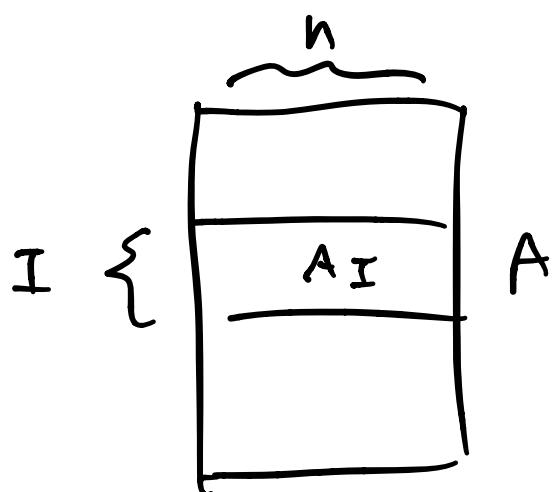
(else $a_i^T x$ for $i \in \bar{I}$ doesn't determine $a_j^T x$, so

$a_j^T x \leq b_j + 1$ has solution

in \mathbb{F} , contradicting $a_j^T x = b_j$,

* Equivalently: submatrix $A_{\bar{I}}$ w/
rows in \bar{I} satisfies

$$\text{row}(A_{\bar{I}}) = \text{row}(A);$$



hence $\text{rank}(A_{\bar{I}}) = \text{rank}(A) = n$.

* Thus: $a_i^T x = b_i$ for $i \in \bar{I}$

has unique soln, so \mathbb{F} is

single point, i.e. a vertex.

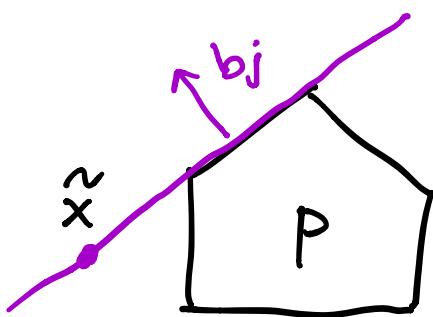
(b) Some inequality needed:

- $\exists j \notin I$ s.t. $\exists \tilde{x}$ w/

$$a_i^T \tilde{x} = b_i; i \in I$$

$$a_j^T \tilde{x} > b_j$$

(b/c $\tilde{x} \notin P$).



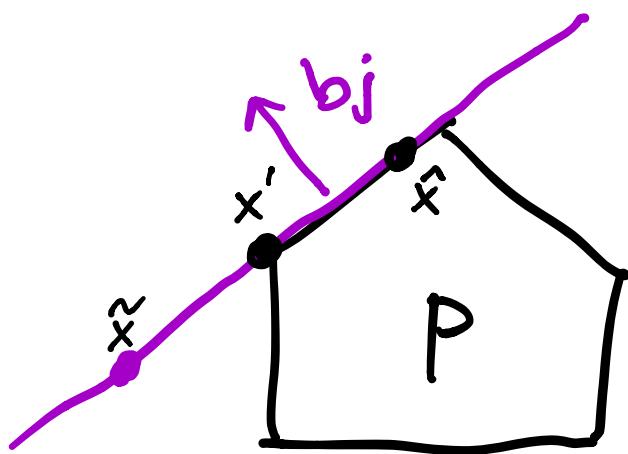
- F nontrivial $\Rightarrow \exists \hat{x} \in F$.

\hat{x} satisfies $a_i^T \hat{x} = b_i$

$$a_j^T \hat{x} \leq b_j.$$

• Consider convex combination

$$x' = \lambda \hat{x} + (1-\lambda) \tilde{x}.$$



• x' satisfies one more equality

(else can move further to x') \square

Finally we can show
equiv b/w bounded polyhedra &
convex hulls. (polytopes).

Recall: $P = \{Ax \leq b\}$ bounded
then $P = \text{conv}(X \text{ extreme pts. of } P)$.
 !!
 X.

Proof: Use TOTD.
 Use TOTD.

- $X \subseteq P \Rightarrow \text{conv}(X) \subseteq P$.
- Assume for contradiction that $\text{conv}(X) \not\subseteq P$.
- Let $\tilde{x} \in P \setminus \text{conv}(X)$.
- Then

$$\sum_{v \in X} \lambda_v v = \tilde{x}$$

$$\sum_v \lambda_v = 1$$

$v \in X$
 $\lambda_v \geq 0 \quad \forall v \in X$
 has no solution.

- TOTA \Rightarrow

$$A = \begin{bmatrix} \text{v's} \\ -I \end{bmatrix} \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$

b

has no soln \Leftrightarrow

$$A^T y = 0, b^T y < 0, y \neq 0$$

has soln. i.e.

$$\begin{matrix} A^T \\ I \end{matrix} \quad y$$

$$\begin{array}{c}
 \left[\begin{array}{ccc|c} -v^T & 1 & & -I \\ \vdots & | & & \\ -v^T & 1 & I & -I \\ \vdots & | & | & \\ \end{array} \right] \xrightarrow{\quad} \left[\begin{array}{c|cc} c & t & s \\ \hline & & \end{array} \right] = \odot \quad (*) \\
 \end{array}$$

for $s \geq 0$, and $\underbrace{y^T b}_{(**)} < 0$.

OR

$$(*) \quad t + c \cdot v \geq 0 \quad \forall v \in X$$

$$(**) \quad t + c \cdot \tilde{x} < 0.$$

- P bounded \Rightarrow

$$\min \{ c^T x : x \in P \} = z^* > -\infty.$$

- Face induced by $C^T x \geq z^*$ nonempty, but contains no vertex.

(because *, ** \Rightarrow objective less on \tilde{x} than any vertex.)

- Contradicts vertex minimality!

(it applies b/c $\text{rank } A = n$,
else $\exists y \neq 0 \text{ s.t. } Ay = 0$,
contradicting boundedness of P .)

